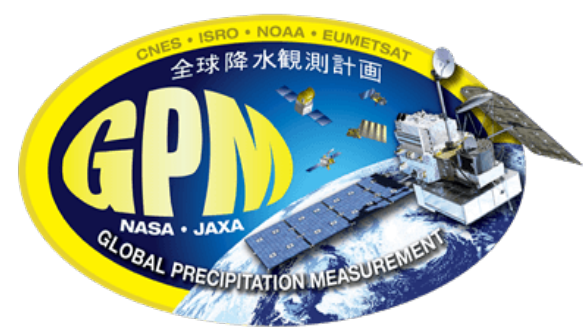




Evaluation of S-band radar rain rate retrieval algorithms and precipitation variability over a dense rain gauge network



David A. Marks^{1,2}, David B. Wolff², Walter A. Petersen³, Pierre –E. Kirstetter⁴, Ali Tokay^{5,6}, Jason L. Pippitt^{1,6}, Jianxin Wang^{1,6}, and Charanjit S. Pabla^{1,2}

¹Science Systems and Applications, Inc, Lanham, MD ²NASA Goddard Space Flight Center - Wallops Flight Facility, Wallops Island, VA ³NASA Marshal Space Flight Center, Huntsville, AL

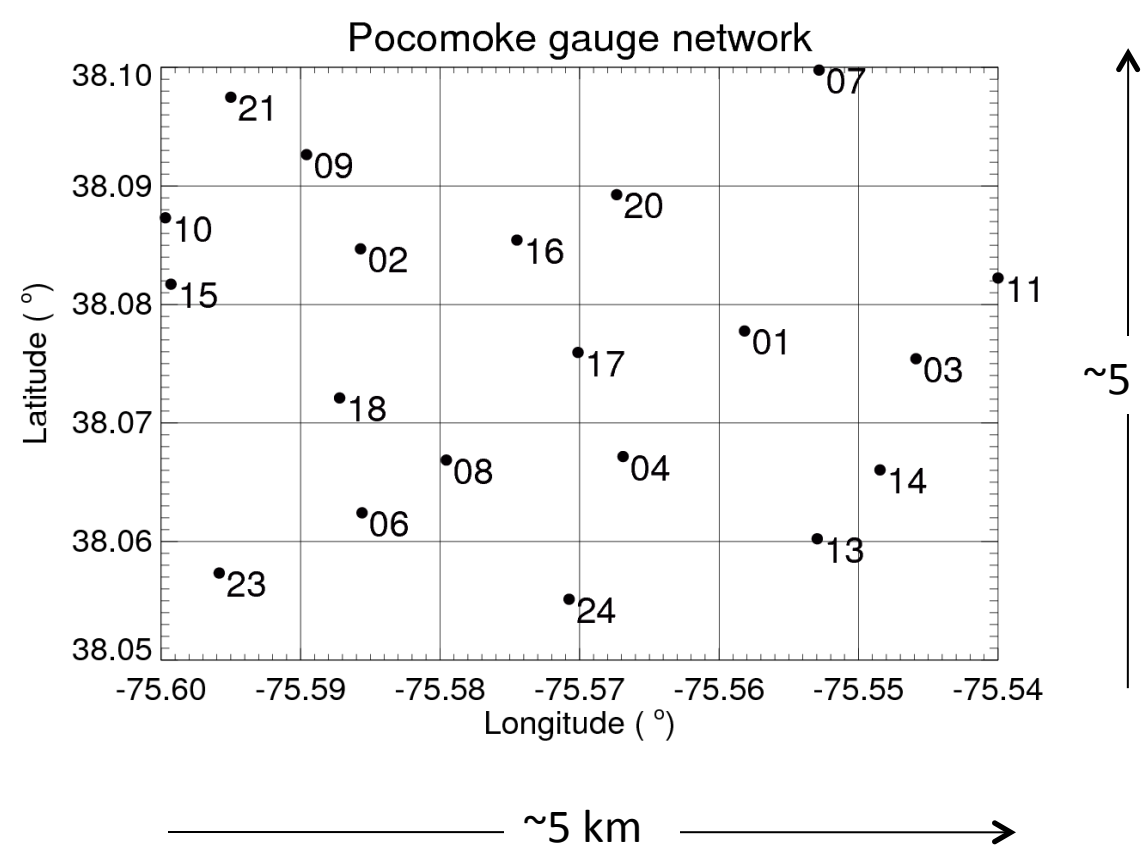
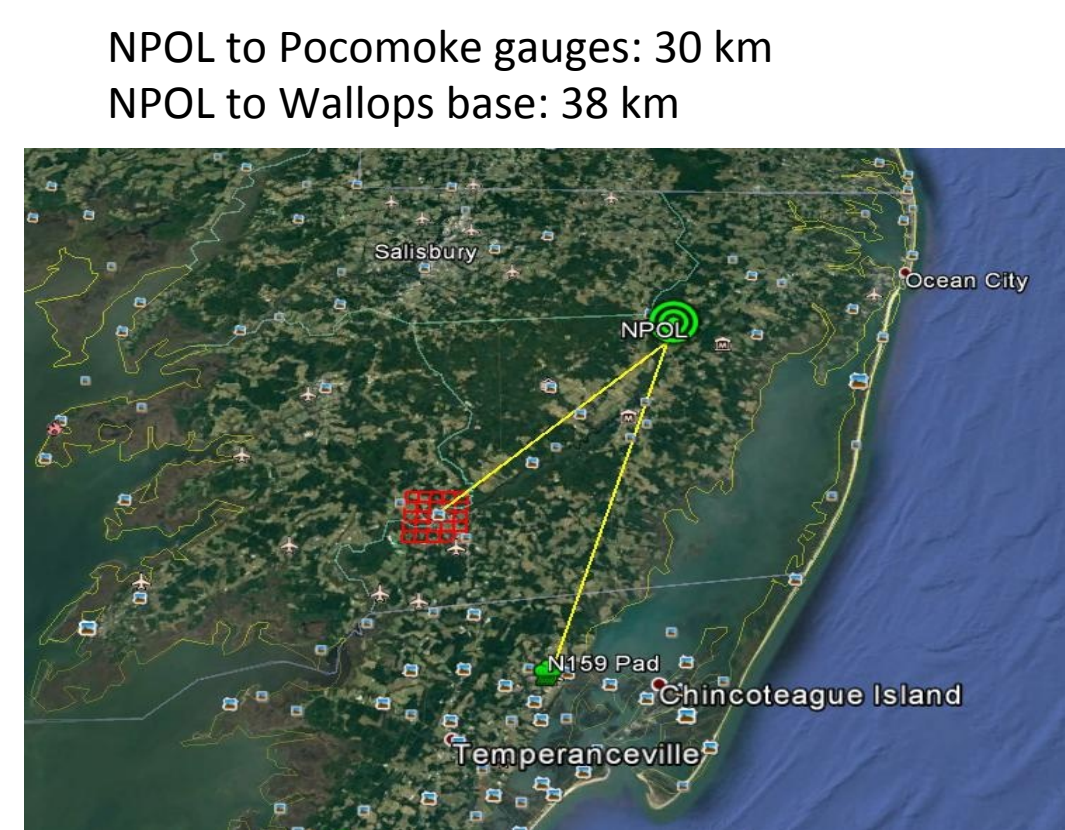
⁴Advanced Radar Research Center, University of Oklahoma / National Severe Storms Laboratory, Norman, OK ⁵Joint Center for Earth Systems Technology, University of Maryland, Baltimore County ⁶NASA Goddard Space Flight Center – Greenbelt, MD

1. Introduction

The Global Precipitation Measurement (GPM) Precipitation Research Facility (PRF) at NASA Wallops Flight Facility has recorded high temporal (50 second) and spatial (250 m) resolution PPI radar data over a dense rain gauge network using NASA's research-quality NPOL (S-band, dual-polarization) radar. The rain gauge network contains 20 tipping bucket gauge pairs distributed through an approximate 25 km² grid located 30km from the NPOL site. Precipitation rates derived from three polarimetric retrieval algorithms (in polar space) were interpolated to a 1.0 km horizontal resolution grid directly over the gauge network. Rain accumulation bias and Mean Absolute Difference statistics from the polarimetric retrievals and the non-polarimetric Multi-Radar/ Multi-Sensor (MRMS) System gauge-adjusted Z-R retrieval (at native resolution of approx. 1km x 1km) were determined via independent gauge comparison from four cases individually and collectively. **The analysis investigates how the statistics from the polarimetric and MRMS retrievals vary from event-to-event and in total over 6, 10, and 14 minute accumulation windows, and if there is a preferred retrieval most appropriate for a specific event type.** In addition, the dimensions of the dense gauge network were intentionally set to be nearly identical to the GPM Dual-frequency Precipitation Radar (DPR) nadir footprint-scale of 25 km². **The rainfall accumulations within the sub-grid scale footprint indicate variability from 100% - 400% depending on event – this is a significant contributor to error within the comparison method.**

2. Instrument / data descriptions and locations

NPOL Characteristics for this study	
Frequency	2.8 GHz
Pulse Repetition Frequency (PRF)	1100 Hz
Wavelength	10.67 cm
Gate Spacing	250 m
Beam Width	0.95 deg (H and V)
Processor	Sigmat RVP-9 / RCP-8



Data descriptions:
*NPOL “Rapid-Scan”: 50-second resolution sector PPI; 1.2 deg elev; Gridded to 1 km H-resolution; Pseudo-CAPPI centered on beam.

*MRMS: Tile 8 RQI–filtered and Gauge-adjusted rates with 2-minute resolution; 0.01 deg grid spacing in latitude and longitude. Comparisons using MRMS are being done at the native MRMS resolution (approx 1km x 1km grid spacing).

*Gauge: Met-One tipping bucket; 0.254 mm per tip; 1-second resolution 20 collocated gauge *pairs* (40 gauges total) in the Pocomoke grid prior to quality control. Only “A” or “B” gauges are selected for each case (20 gauges prior to QC).

3. Rain rate estimation algorithms

RR (“DROPS2.0” Chen et al. 2017)

- Incorporates a dual-pol QC algorithm and Kdp estimation.
- “Region-based” hydrometeor ID instead of bin-by-bin.
- Includes RUC model sounding vertical temp profile as input.
- Rain rate relation equations based on DSD observations from 14 APU disdrometers employed in NASA’s IFloods field campaign (Iowa 2013).
- Architecture of DROPS2.0 is similar to Cifelli et al. 2011

RP (Bringi et al. 2004):

- Z-R relation of the form $Z = aR^{1.5}$
- Coefficient “a” continuously adjusted as DSD evolves in space/time.
- Normalized gamma DSD parameters estimated via radar measurements of Zh, Zdr, and Kdp (via Gorgucci et al. 2002)
- Method continuously estimates DSD parameters
 - no classification of rain type is needed.

RC (Cifelli et al. 2003, 2011)

- Polarimetric optimization algorithm driven by hydrometeor ID (HID).
- Our application in this study does not consider mix or ice – only rain.
- Rain rate estimation equations: $[R(Zh), R(Zh,Zdr), R(Kdp, Zdr), R(Kdp)]$ using pol-variable thresholds.
- Pol equations physically based – derived from range of gamma DSD parameters typically found in observations (simulated obs).
- Assumes drop shape following Beard and Chuang (1987) equilibrium model.

MRMS (Zhang et al. 2011, 2016):

- WSR-88D Reflectivity mosaic (3D) with exponential weighting function based on distance and height.
- DP Quality Control + VPR + RQI (based on blockage and beam height).
- Radar-based QPE (R-Z) using automated sfc precip classification + NWP model data
- NOAA Hydrometeorological Automated Data System (HADS) gauges
- Radar-based QPE can be modified via local gauge bias correction (inverse – distance-weighted).

4. Case descriptions and comparison results

Cases (all from 2015)	Event Description	NPOL data	# gauges for NPOL	# gauges for MRMS
June 2	Moderate strat w/conv	2.5 hours	18	11
	Light stratiform	1.5 hours		
June 21 (TS Bill)	Strong conv	1 hour	18	9
	Mod / Heavy strat	6.5 hours		
June 24	Light/mod strat w/minor conv coverage	3.5 hours	18	9
	Strong conv w/ mod strat	1 hours		
June 26	Light/mod strat	5 hour	18	10

Accumulations from all estimates are evaluated over 3 time windows (6, 10, and 14 minutes).

Bias = $(\sum \text{Radar} - \sum \text{Gauge}) / \sum \text{Gauge} \times 100$.

Mean Absolute Difference (MAD) = $\langle |\text{Radar} - \text{Gauge}| \rangle / \langle \text{Gauge} \rangle \times 100$.

R = correlation coefficient

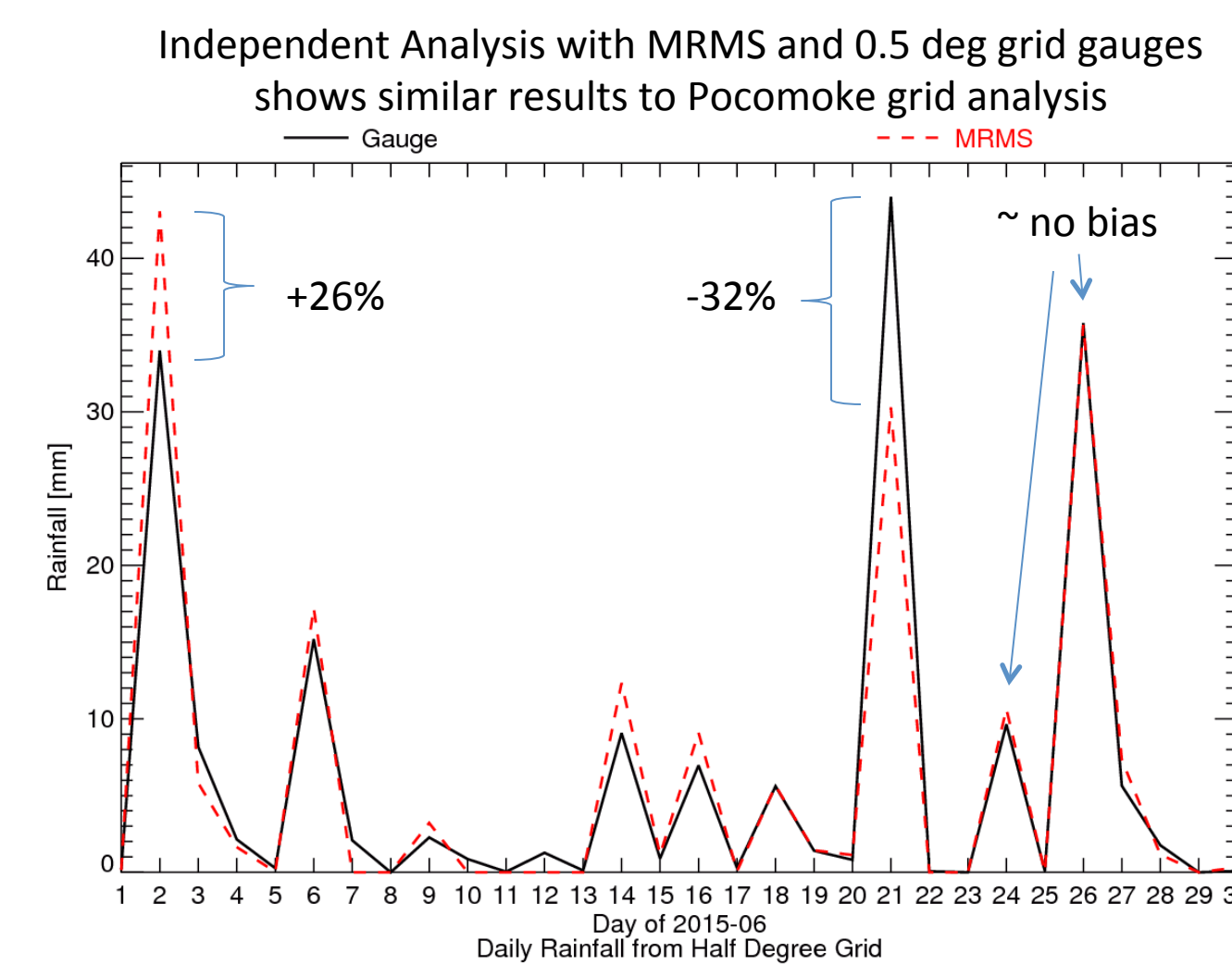
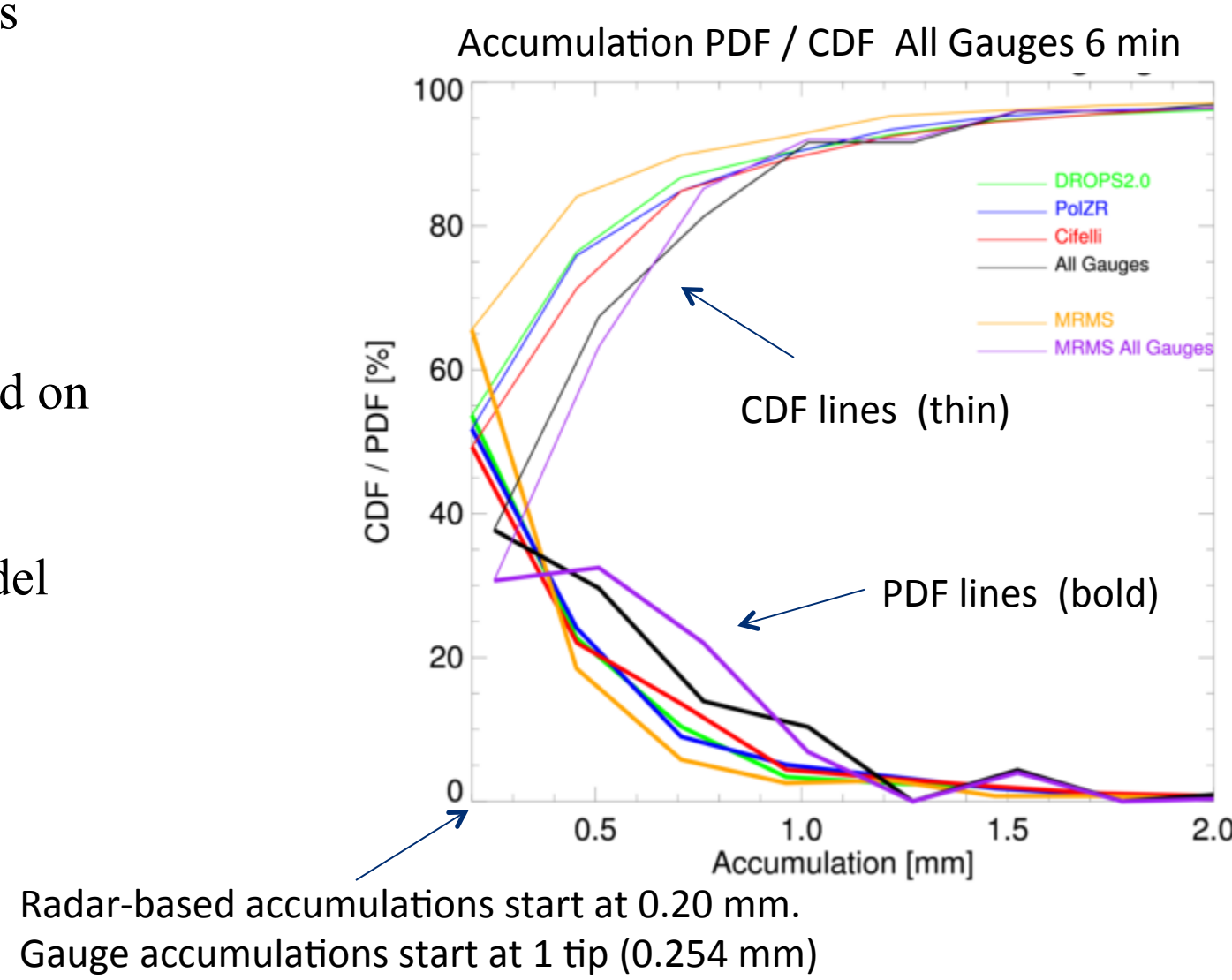
June 2	Bias [%]			MAD [%]			R			Samples		
	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m
RR (DROPS2.0)	3.1	2.9	8.8	24.5	21.2	25.1	0.93	0.94	0.93	165	129	109
RP (Bringi)	-9.5	-12.0	-3.4	23.2	22.7	24.1	0.94	0.94	0.93	158	120	105
RC (Cifelli)	16.1	16.9	27.8	27.5	23.6	31.9	0.94	0.96	0.94	178	132	109
MRMS (NOAA)	17.2	20.9	31.0	25.7	30.4	36.0	0.95	0.92	0.92	92	69	58

June 21	Bias [%]			MAD [%]			R			Samples		
	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m
RR (DROPS2.0)	-17.1	-24.3	-28.1	24.0	29.3	32.5	0.93	0.89	0.90	281	247	218
RP (Bringi)	-5.9	-4.7	-5.2	14.7	13.1	11.7	0.97	0.98	0.98	449	329	277
RC (Cifelli)	-10.0	-11.7	-13.5	26.2	27.4	28.0	0.89	0.90	0.91	356	288	235
MRMS (NOAA)	-37.5	-42.8	-39.3	40.8	45.4	44.0	0.77	0.80	0.84	190	190	167

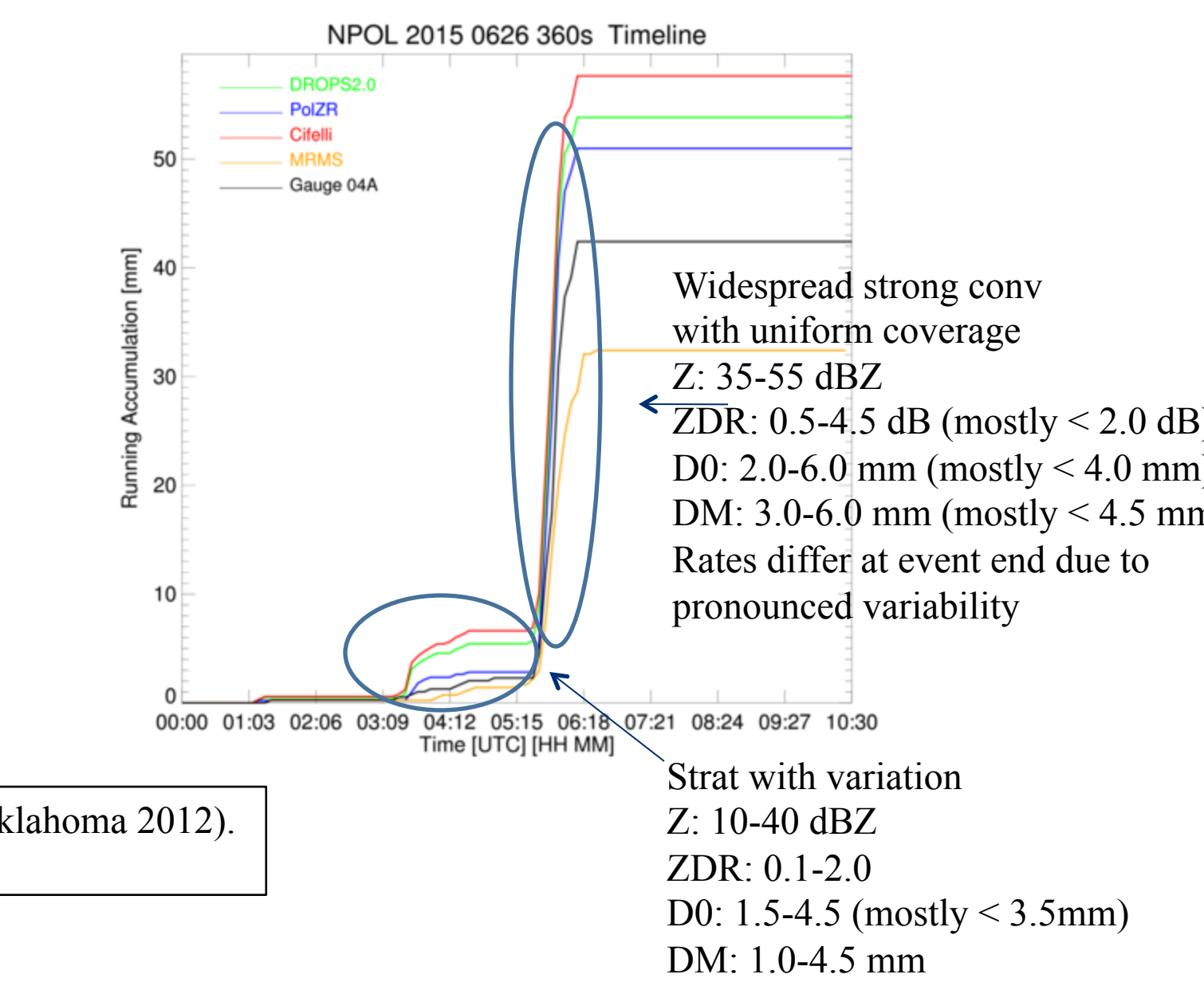
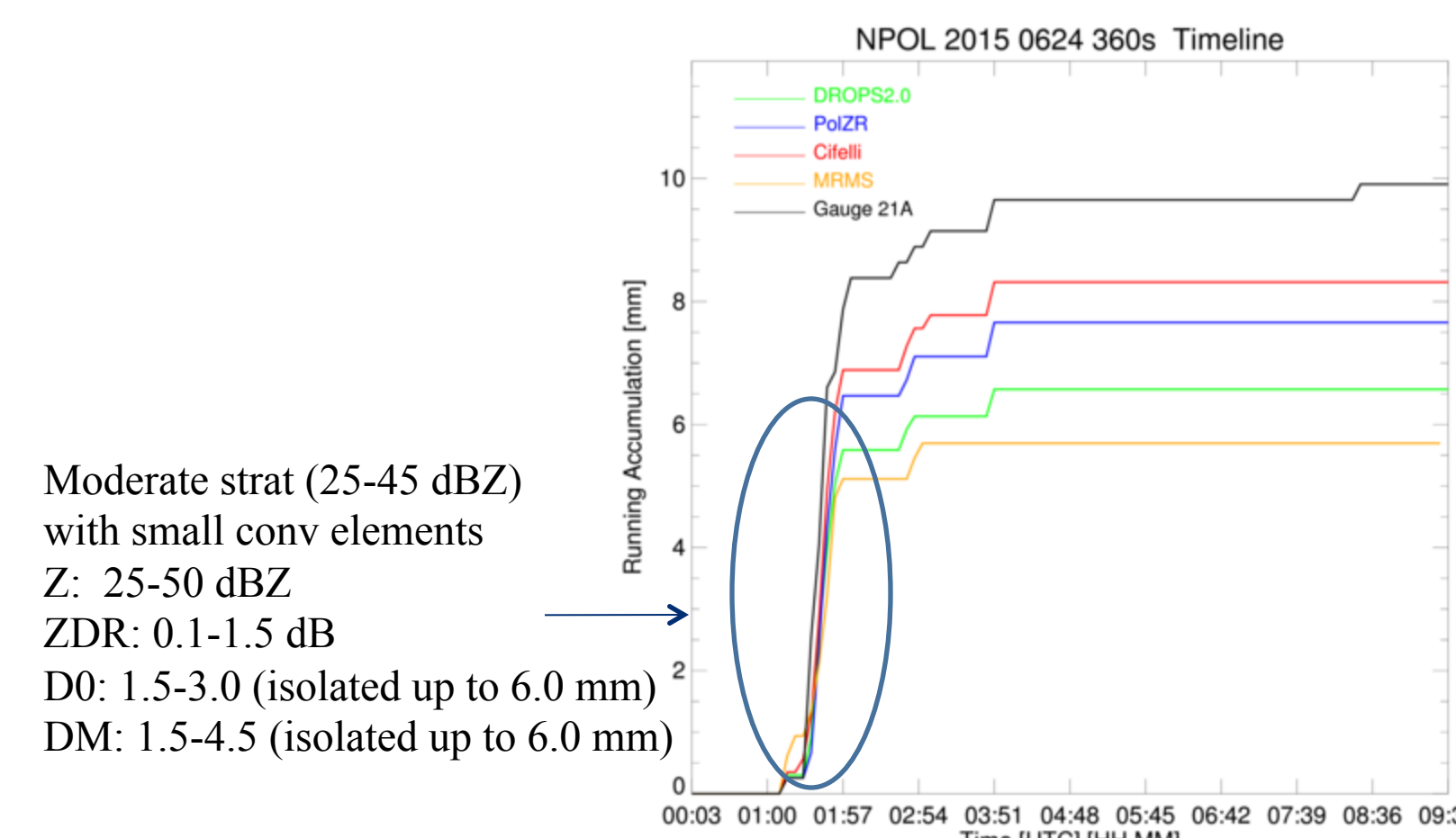
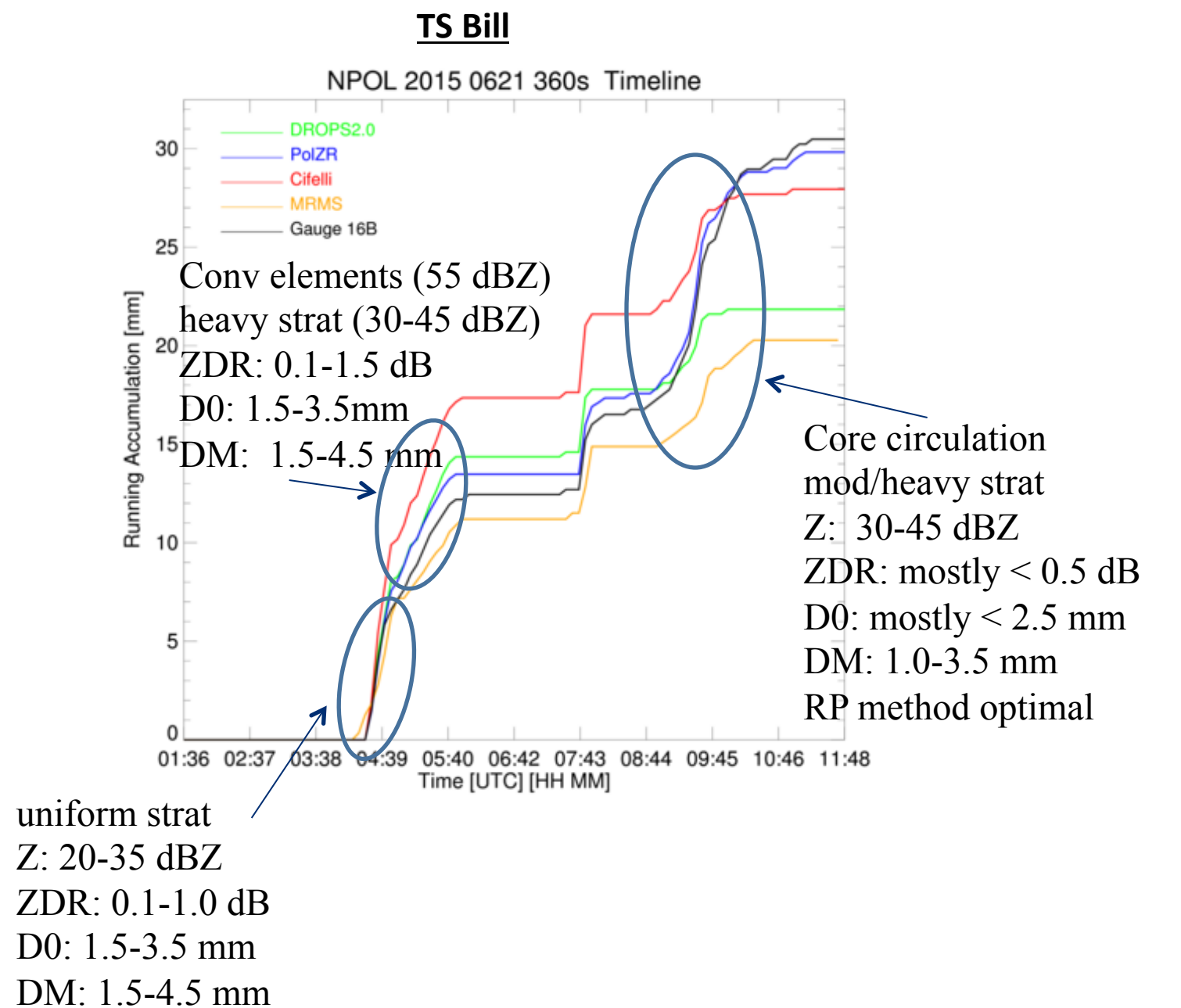
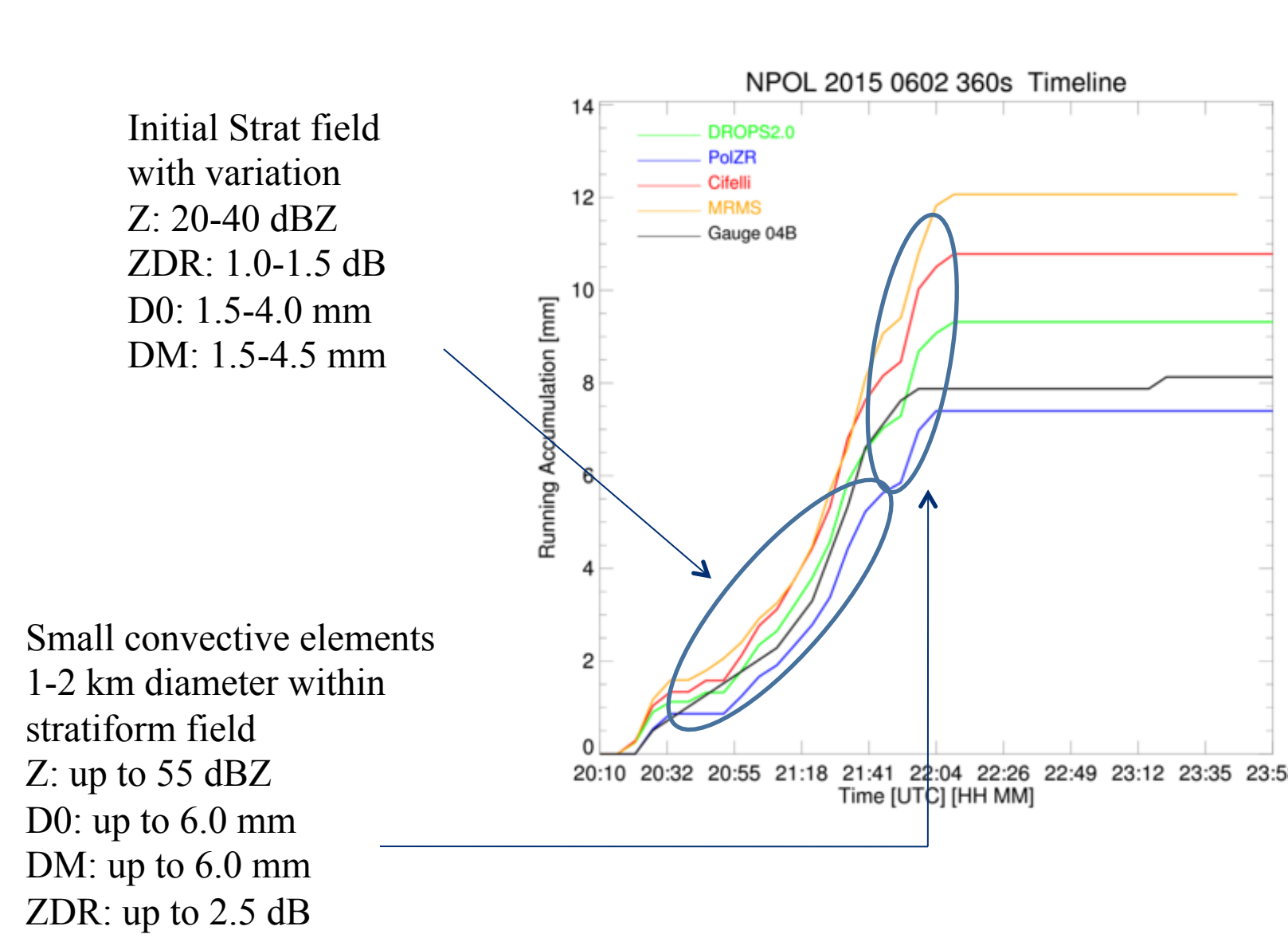
June 24	Bias [%]			MAD [%]			R			Samples		
	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m
RR (DROPS2.0)	-7.4	-6.1	-7.4	11.1	8.1	11.2	0.99	0.99	0.99	25	27	24
RP (Bringi)	-3.5	-6.4	-4.0	8.0	7.7	6.7	0.99	0.99	0.99	27	27	30
RC (Cifelli)	-1.1	-1.8	0.9	7.1	8.9	8.3	0.99	0.99	0.99	26	30	32
MRMS (NOAA)	-2.7	-11.2	-1.2	11.8	22.1	18.0	0.97	0.97	0.99	7	12	14

June 26	Bias [%]			MAD [%]			R			Samples		
	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m
RR (DROPS2.0)	7.6	11.8	24.2	8.8	12.3	25.1	0.99	0.99	0.99	119	128	105
RP (Bringi)	4.5	7.5	12.6	7.2	10.0	14.9	0.99	0.99	0.99	87	106	97
RC (Cifelli)	12.7	23.9	27.9	13.5	23.9	28.4	0.99	0.99	0.99	138	133	116
MRMS (NOAA)	-2.1	-3.1	-1.3	9.3	22.3	21.7	0.99	0.96	0.99	24	65	67

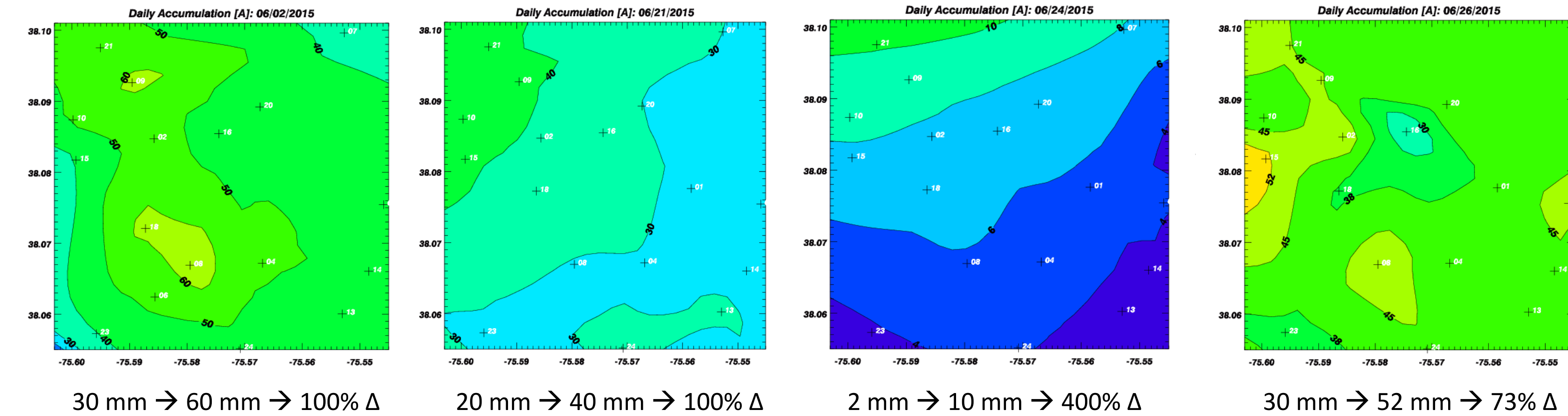
All cases	Bias [%]			MAD [%]			R			Samples		
	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m	6 m	10 m	14 m
RR (DROPS2.0)	-3.4	-3.3	-2.1	14.9	16.3	21.2	0.99	0.99	0.98	559	508	423
RP (Bringi)	-2.4	-2.0	-0.8	12.5	12.9	11.7	0.99	0.99	0.99	668	565	482
RC (Cifelli)	5.1	8.9	12.4	16.5	17.7	20.8	0.99	0.99	0.99	636	536	456
MRMS (NOAA)	-16.8	-21.6	-16.6	24.1	30.4	29.7	0.97	0.91	0.93	277	315	286



5. Algorithm differences through timeline accumulations



6. Rainfall accumulation variability within a DPR 5km x 5km pixel



7. Observations

- Both gauge and radar quality control are critical to comparisons.
- The rain rate and accumulation variability over the 25 km² grid are significant contributors to error .
- There is no clear-cut optimal choice in radar-based estimation algorithms (except for tropical system).
- For a modified tropical event, rate estimation using continuously adjusted DSD parameters (Bringi (RP) June 21 event) is clearly optimal compared to the blended polarimetric algorithms and MRMS.
- The correlation does not show a significant difference between the 6, 10, and 14 minute accumulation periods.
- MAD for polarimetric estimates is 5-10% lower than MRMS with all events combined.
- Within stratiform, increased variability in Z, ZDR, and D0 / DM lead to divergence in the polarimetric estimates over time (especially with numerous embedded convective elements).
- A stratiform (or convective) event with minimal variability in Z, ZDR, and D0 / DM lead to similar rates and accumulations from the polarimetric estimates and are highly correlated with gauges.
- The Cifelli (RC) algorithm generally overestimates rates relative to the other polarimetric and MRMS algorithms, but often has the highest correlation with rain gauges.
- The DROPS2.0 (RR) algorithm produces rates lower than RC due to revised default Z-R relation and region-based HID.
- Future work: sensitivity testing of accumulation thresholds from 0.25 mm to 1.0 mm.

Acknowledgements:

Dr. Ramesh Kakar, NASA HQ
Dr. Gail Skofronick Jackson, GSFC: GPM Project Scientist and Mesoscale Atmospheric Processes Lab Chief: funding and support
Dr. Scott Braun, GSFC, TRMM Project Scientist: funding and support
Michael Watson and Gary King: NPOL radar engineers
Wallops PRF ground instrument support team
Dr. V. Chandrasekar and Haonan Chen (CSU): DROPS2.0 executable code